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Business Background



Challenges

- Custom Retention
- Annual Savings



AI Model

- Lacks Model
- Gap in Fraud Prevention



Main Takeaway

Enhance:

- Customer Retention
- Promote Annual Savings

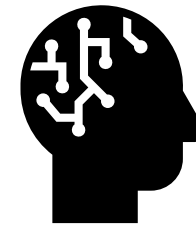
Fraud

- Increase of 149%
- Increase of 83%
- \$56 Billion Lose



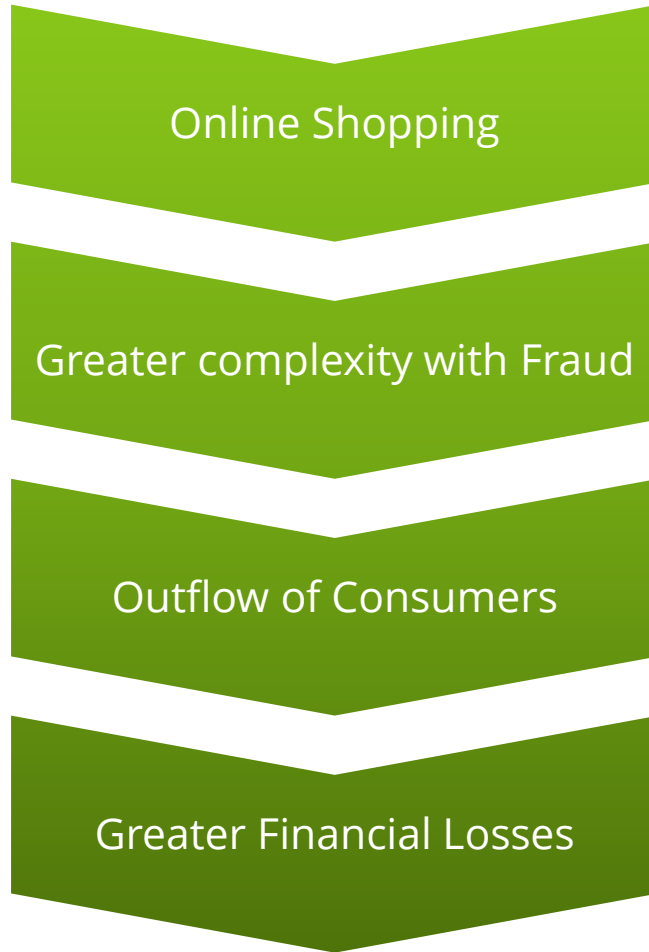
AI

- 300-Billion-Dollar-Impact
- Positive ROI
- Balance Technology & Stakeholder Interests
- Human Intervention



Fraud & AI

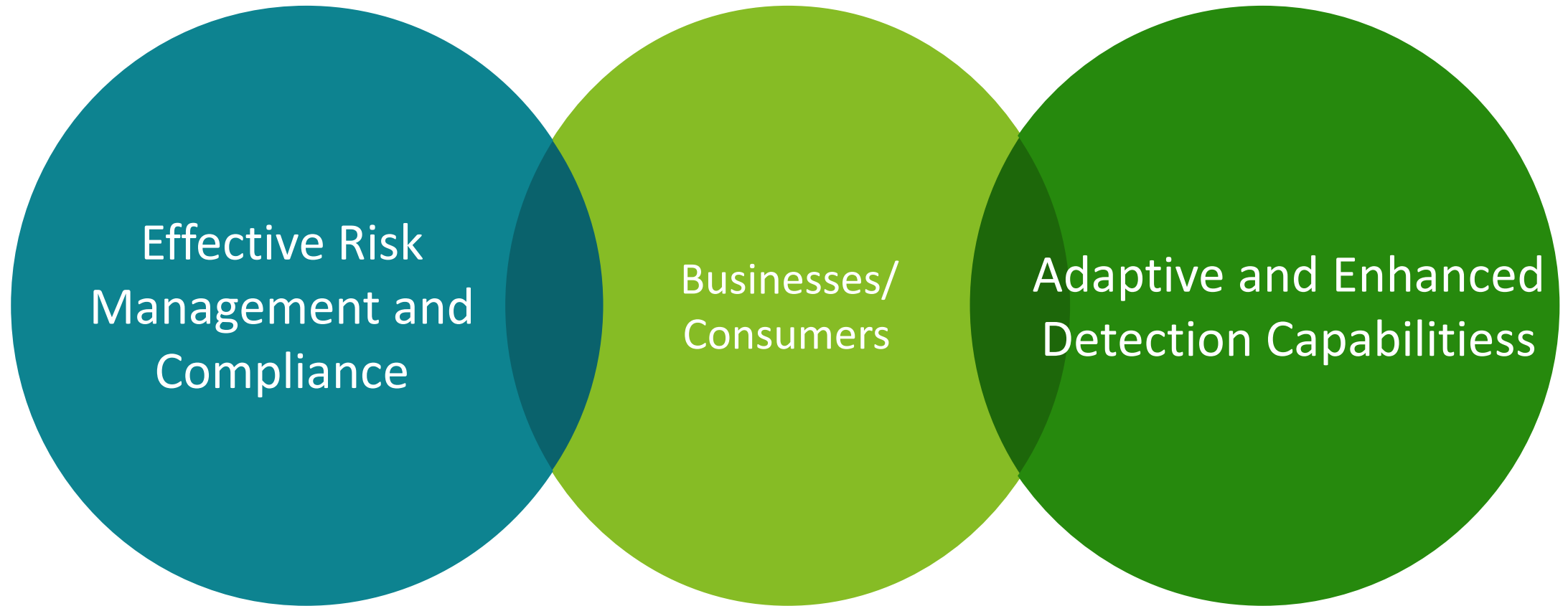
Fraud's Business Impact



- Online scammers
- Additional vectors for fraud
- New trends in fraud
- AI specialized tactics
- Fraud = customer dissatisfaction
- High income/net worth clients
- Growing fraud exposure

Business/Consumer Needs For AI

AI transforms banking fraud detection with real-time monitoring, dynamic security, and instant fraud identification, addressing the need for robust security measures and enhancing both business risk management and compliance, while also prioritizing a positive consumer experience by minimizing the impact of false positives.



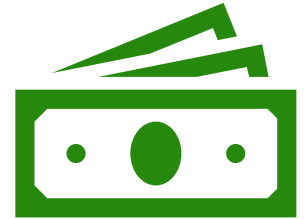
Objectives



Fraud Detection

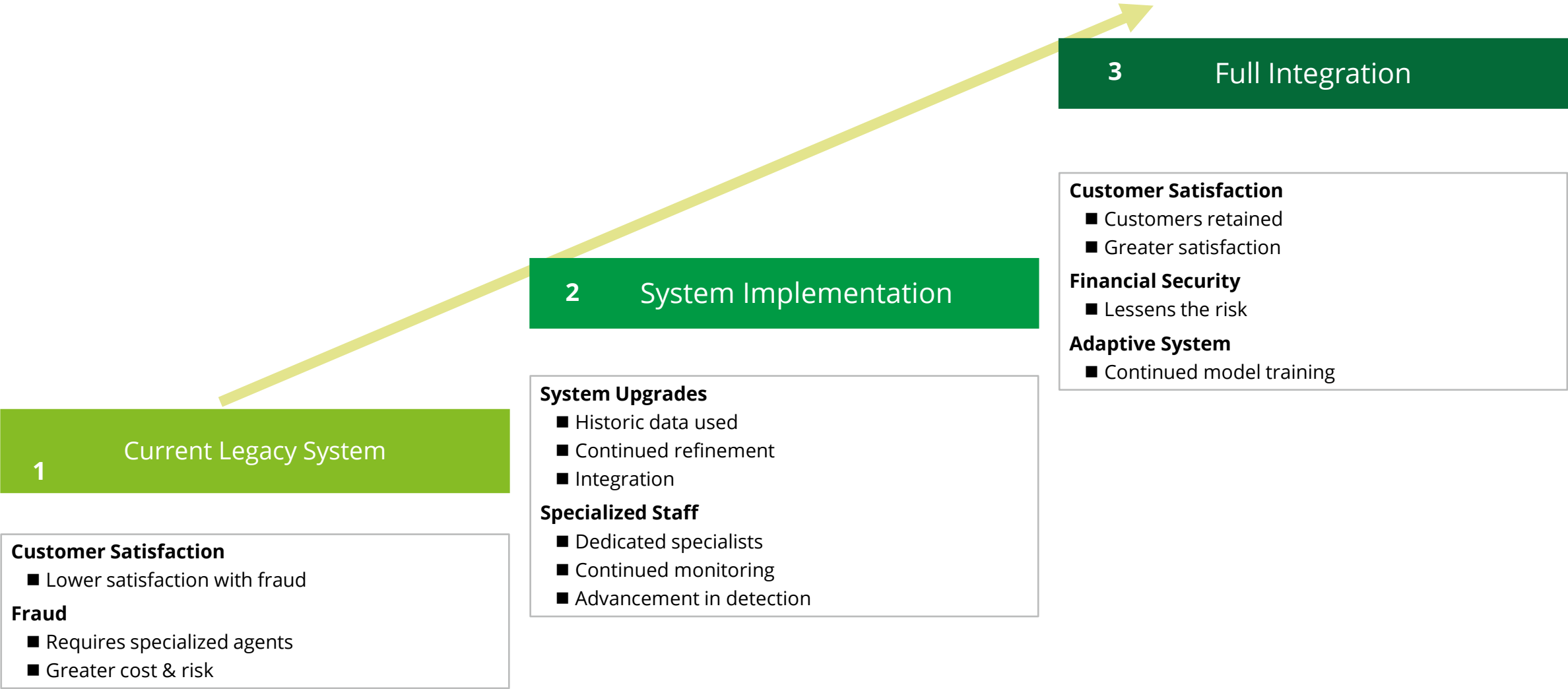


Customer Experience



Financial Security

Timeline of Fraud Detection



Data Analysis

Data Metrics

Distance from home



Distance from the last transaction



Ratio to Median Purchase price



Repeat Retailer



Is Fraud?



Definition



- The distance from home where the transaction happened



- The distance from last transaction happened



- Ratio of purchased price transaction to median purchase price.



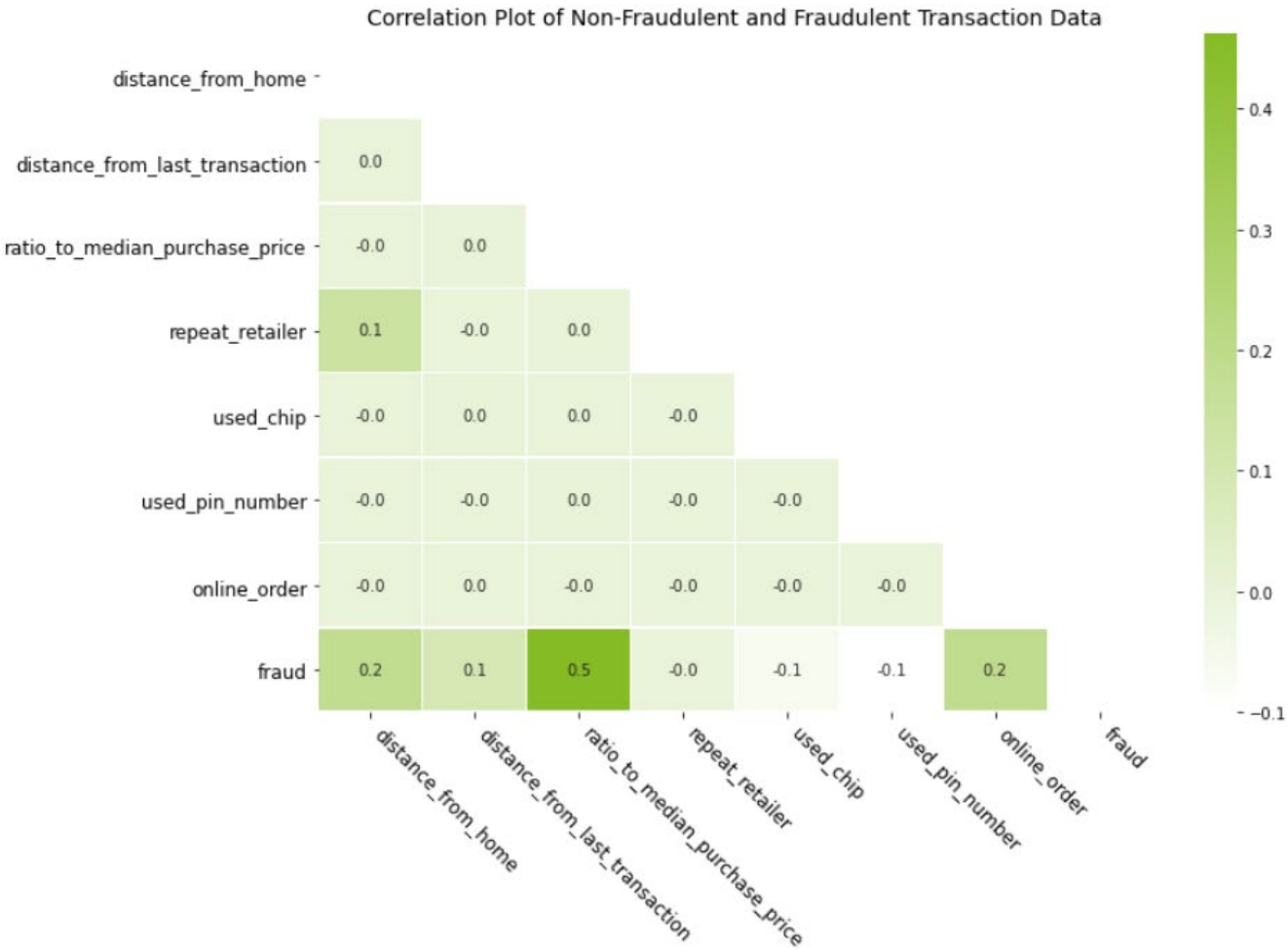
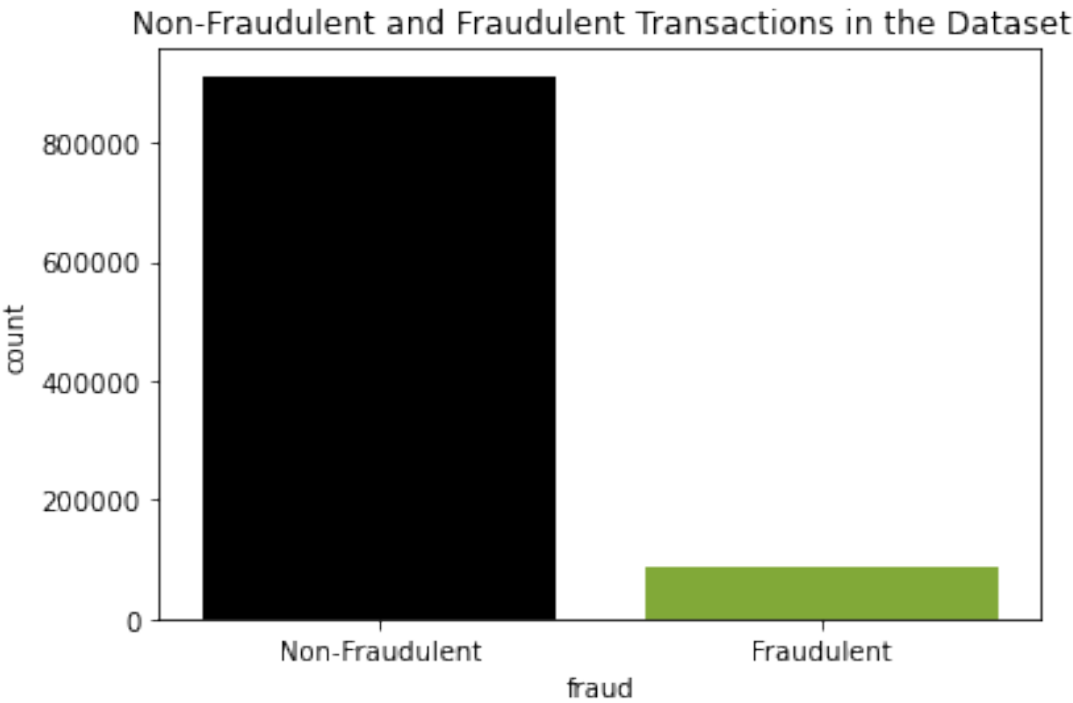
- Is the transaction happened from same retailer



- Is the transaction fraud (1) or not fraud (0)?

Model Analysis

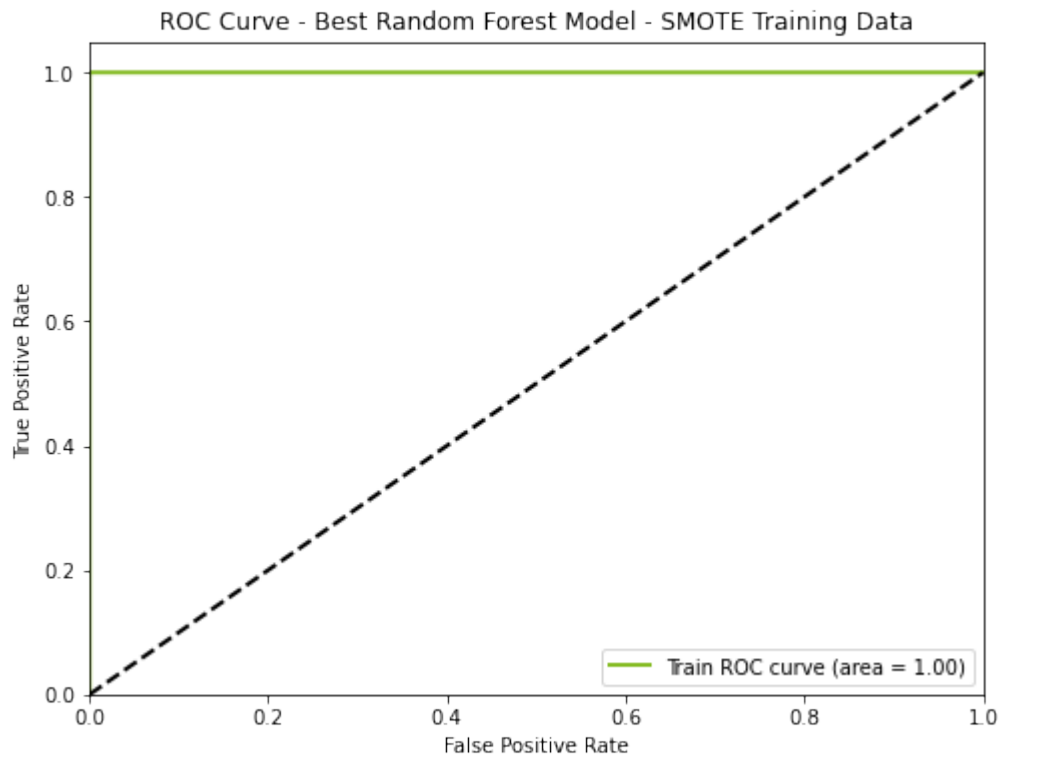
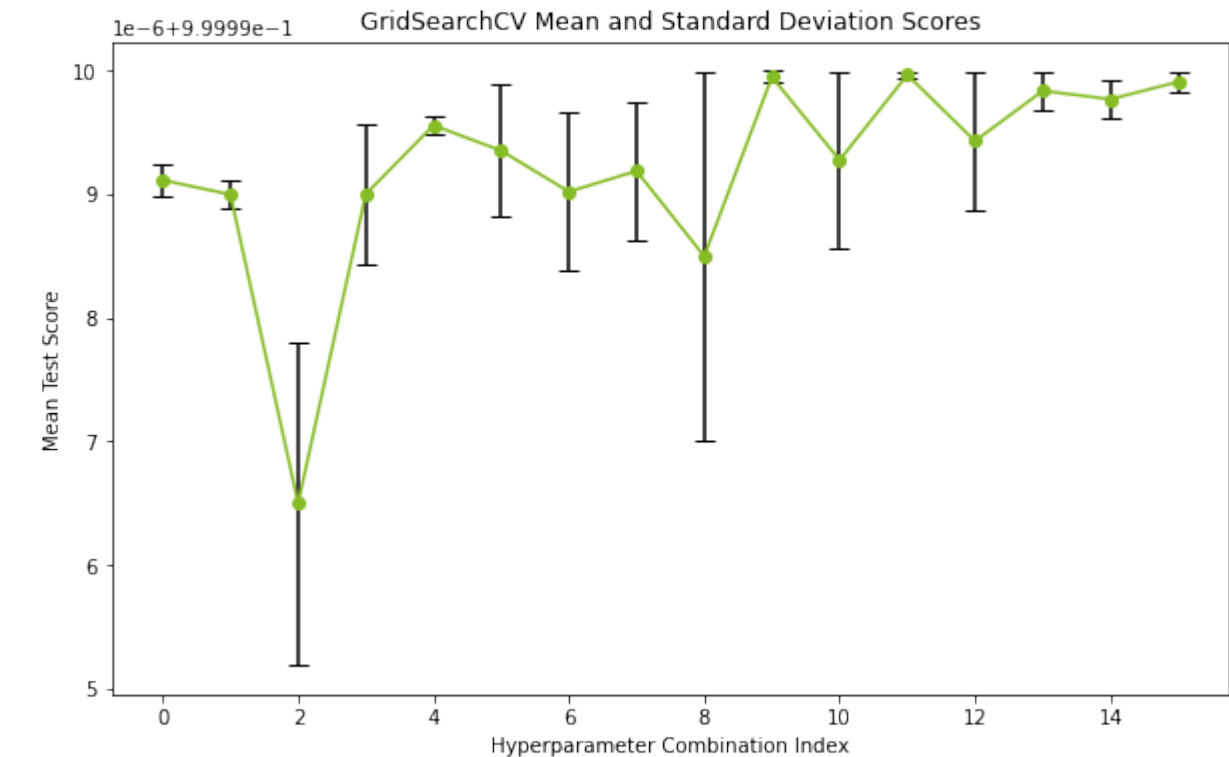
This EDA model displays the relationship of fraud (1) to the other attributes in the data.



Performance Metrics of Model with Balanced Data Using SMOTE and Cross-Validation

The Random Forest model showcases exceptional performance metrics after balancing the data using SMOTE. The GridSearchCV results indicate a successful cross-validation with optimal hyperparameter values selected. Mean cross-validated score: 1.0, Standard deviation of the cross-validated scores: 0.0.

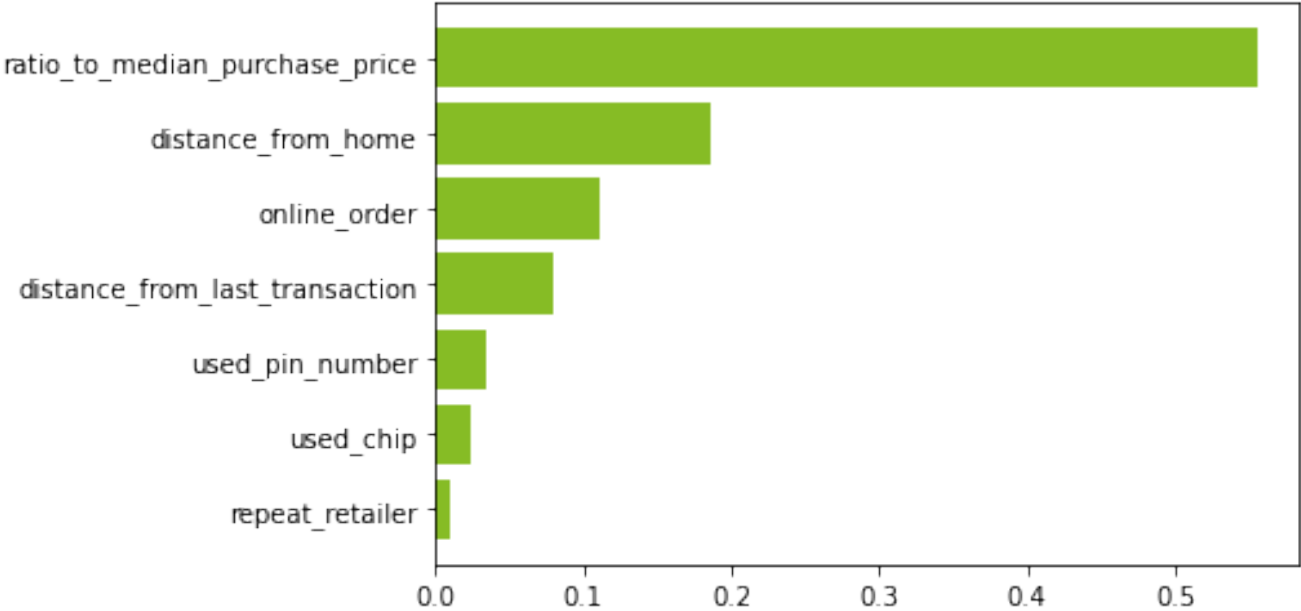
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Random Forest Model with SMOTE Training Data Metrics:  
Accuracy: 0.99997  
Precision: 0.9999427819419809  
Recall: 0.999713975173045  
F1 Score: 0.999828365467132  
ROC AUC: 0.9998542481456972
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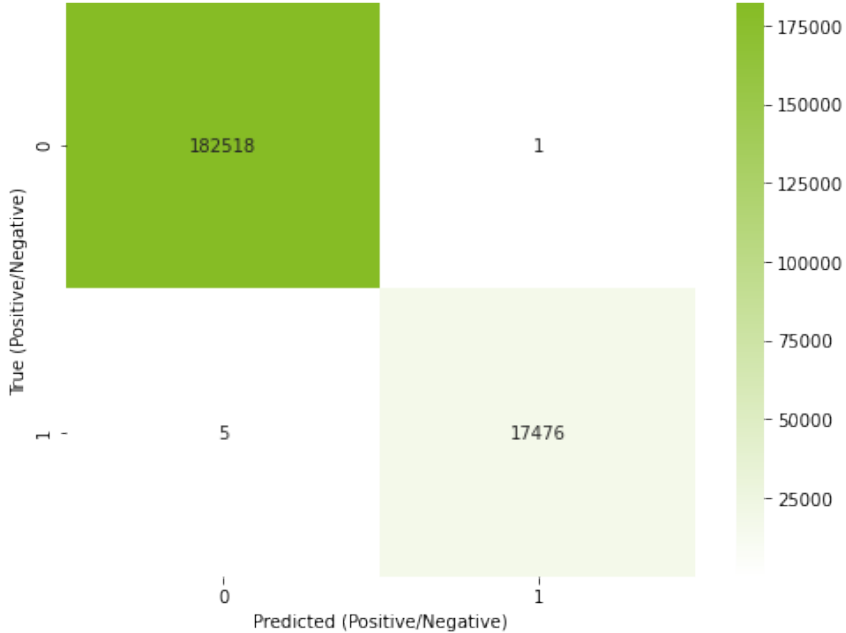
Model Features & Confusion Matrix

This precision driven analysis reveals our model's exceptionally robust performance in safeguarding against fraudulent transactions.

Feature Importances of Random Forest Model (SMOTE-enhanced Data)



Confusion Matrix: Assessing Credit Card Fraud Detection Performance



ROI & Risks



ROI

Reduction in fraud related losses
Increase customer retention



Model Accuracy

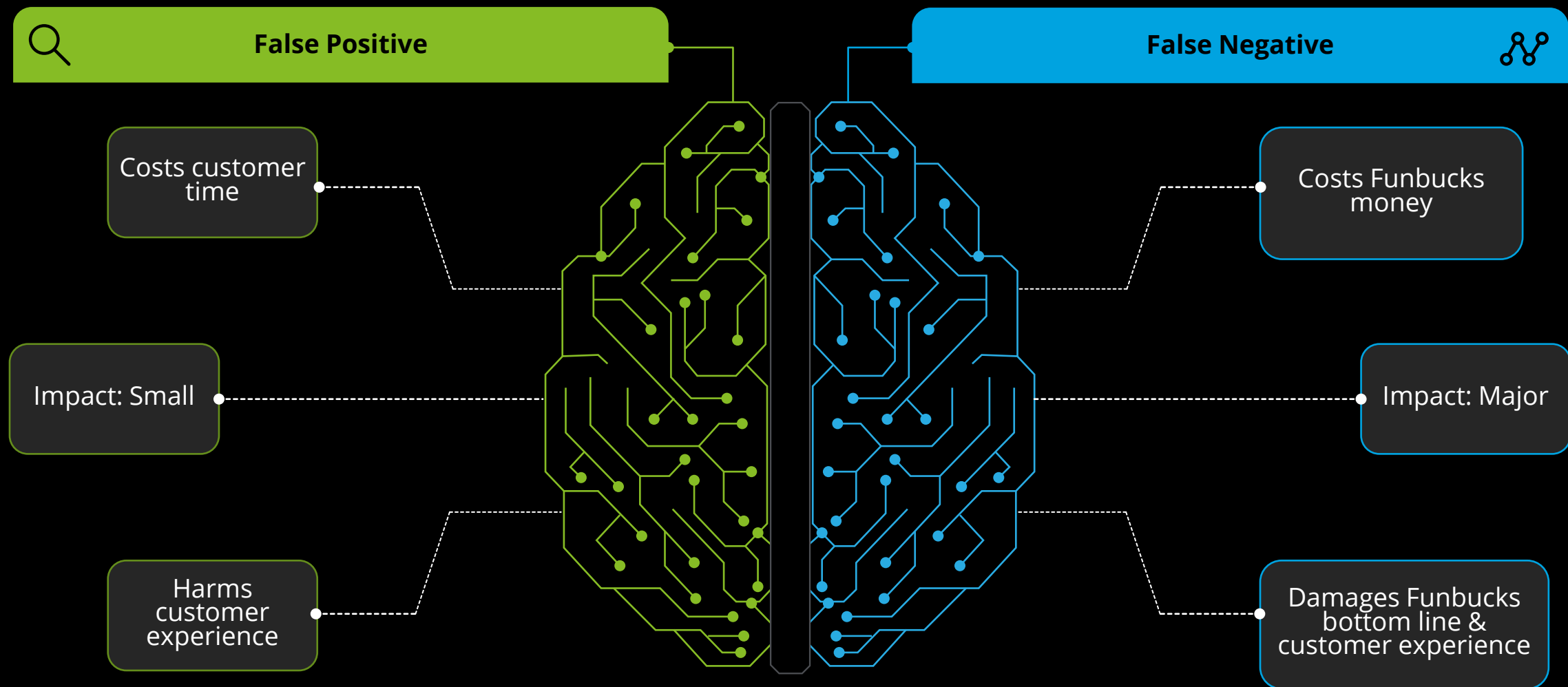
Continued investment in fraud detection
Update based on changing fraud patterns



Regulatory Compliance

Data protection and transparency

False Positives and False Negatives



Conclusion

Reduction in fraud increases operating revenue and increase customer retention

Cost Savings



Leaders in Technology

Become industry leaders in fraud detection solutions



Continued Adaptation

Remain ahead with continued refinement

Deloitte.

Thank You! Questions?

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